

# Background on Adaptive Management and Decision Support Systems



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# Decision support tools

Ideally an easy-to-use quantitative tool for evaluating tradeoffs

A tool that can:

- Incorporate monitoring data
- Involve scientists working within TRRP
- Allow decision makers to evaluate tradeoffs
- Identify key uncertainties that make decisions difficult



# Decision support tools

Integrated decision support tools are most effective when:

- 1) Decisions are revisited regularly
- 2) Decisions can be informed by ongoing monitoring
- 3) **Managers and scientists take ownership of the process**



# Decision analysis-- structured decision making-- adaptive resource management--- decision support tools– decision support systems

## Some background

Decision theoretic approaches

- Developed turn of 20<sup>th</sup> century

- Initially business applications

- 1970's natural resource decision making

Adaptive resource management

- Special case of DA

- Feedback from monitoring



# Decision Support Process

Requires thought (time to think)

Break down problem into manageable parts

Management objectives

- what do you really want to accomplish?

Alternative decisions/ actions

Link decisions with objectives

- a model

Evaluate sensitivity of decisions to assumptions

identify monitoring endpoints

estimate required effort- evaluate tradeoffs

Link with monitoring (ARM)



# Don't we already do this?

Confusion over objectives (fundamental vs. means)

Confusion between goals/ objectives and beliefs/  
technical uncertainty

Failure to adequately consider sources of  
uncertainty (blind faith in models)

Failure to incorporate research and monitoring into  
decision making (to reduce uncertainty) →  
inefficient use of resources

# Why model decisions?

Quantitative methods should guide and support decision making

Not a replacement for human intuition and subjectivity

At least- be an intelligent consumer of models for decision making

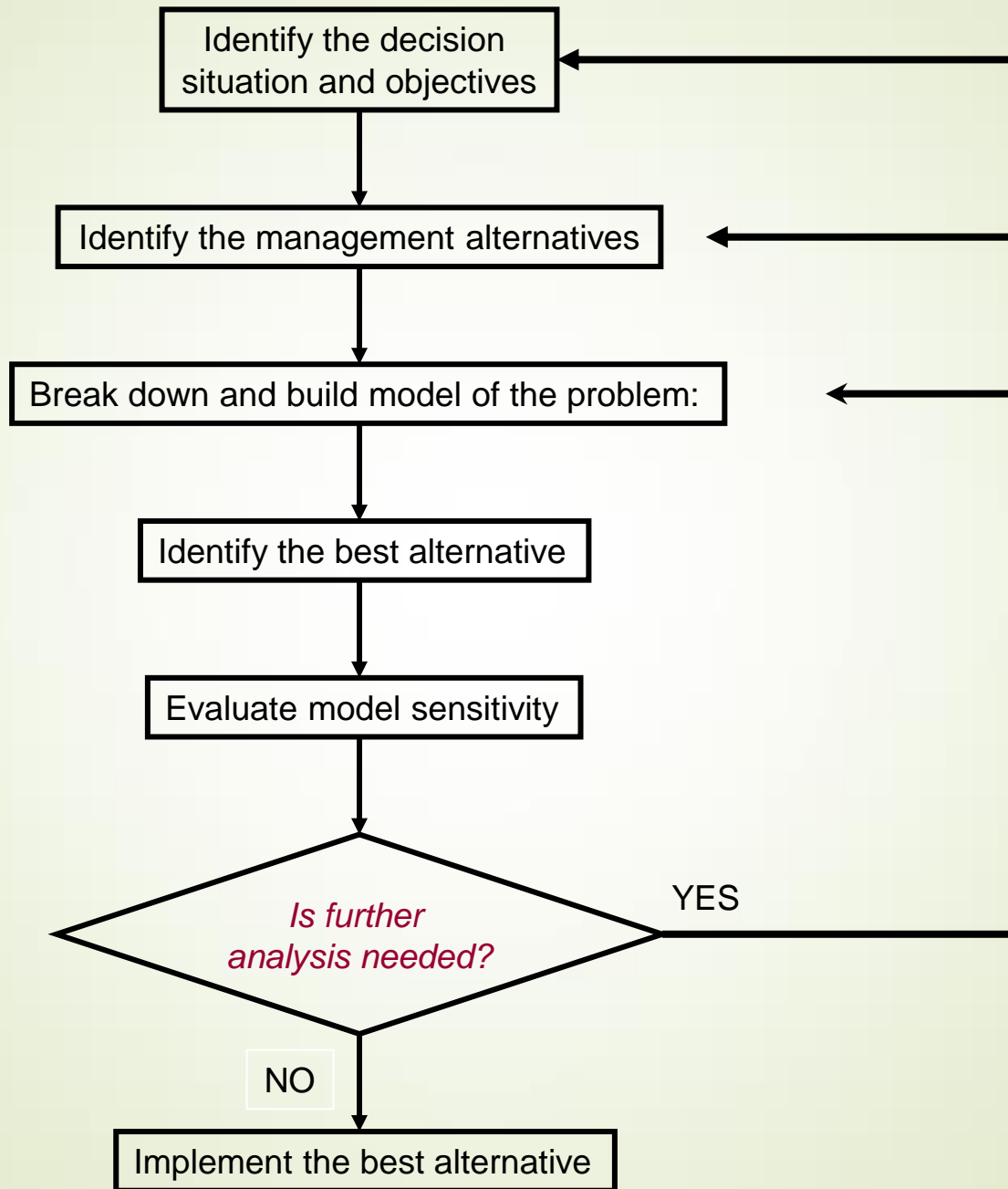
Better: collaborate on model development

Models for conservation are *not* about modeling--  
- they are about conservation

# The Big Picture

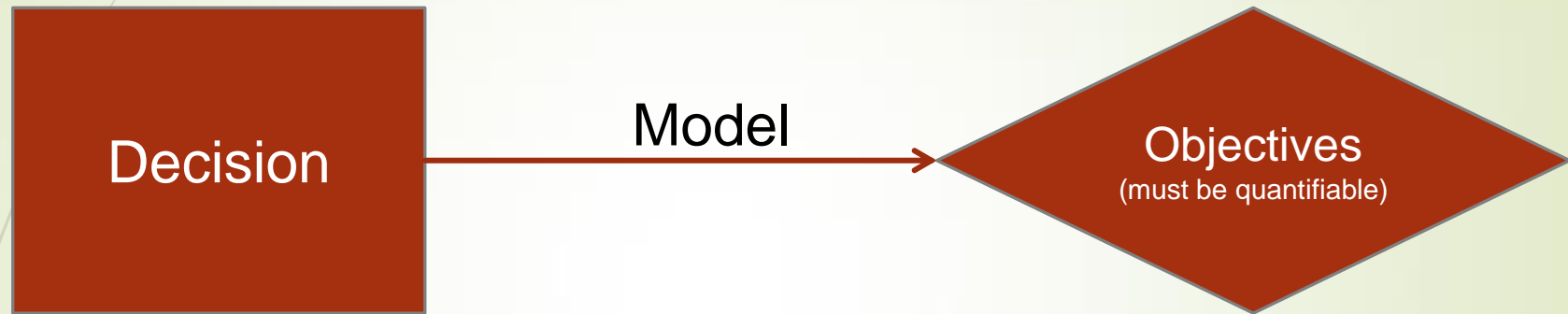


# Decision support process



## Step 2: Identify and Structure Objectives

Why emphasize objectives?



Everything depends on your objectives

Everything depends on your objectives

Everything depends on your objectives

# Basic types of objectives

Fundamental objectives: what the decision-maker really wants to accomplish.

Means objectives: the things that need to be accomplished to realize the fundamental objective

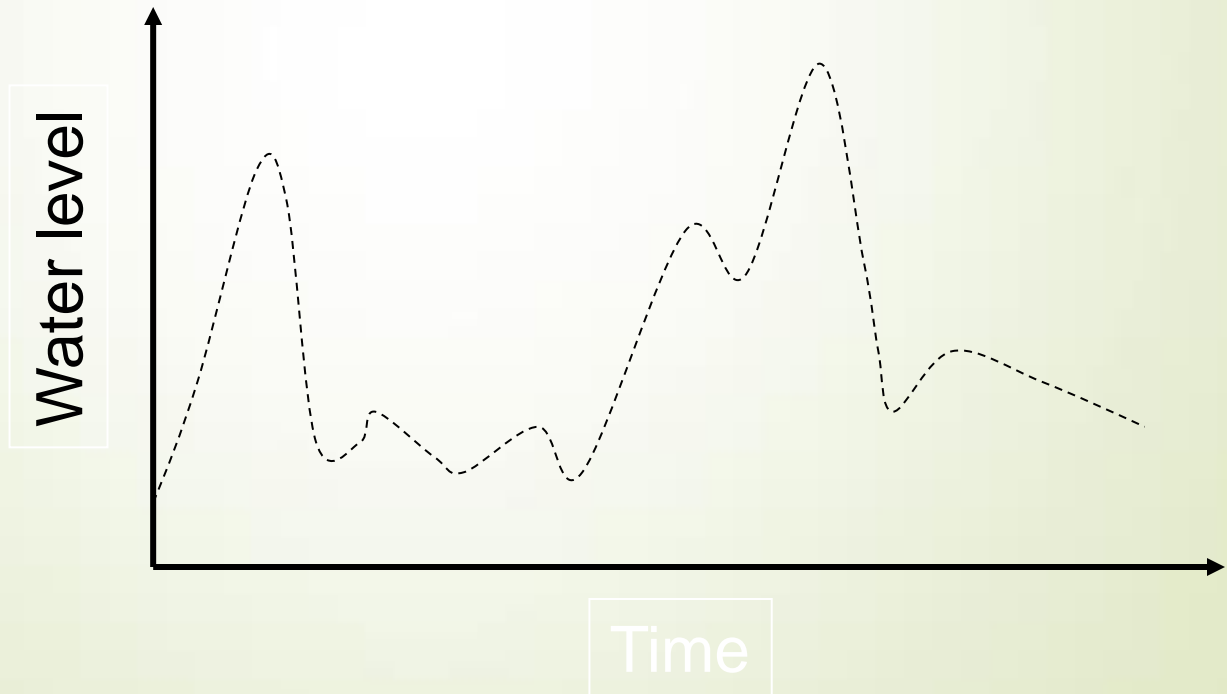
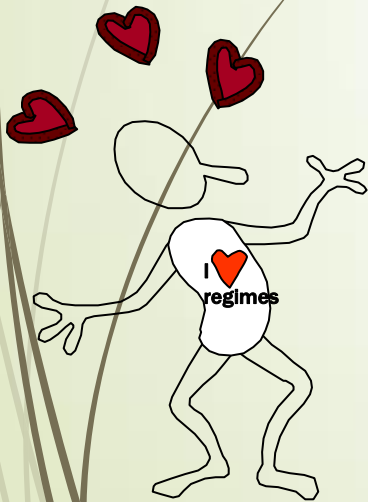


>>>>> Clarity is essential <<<<<<<

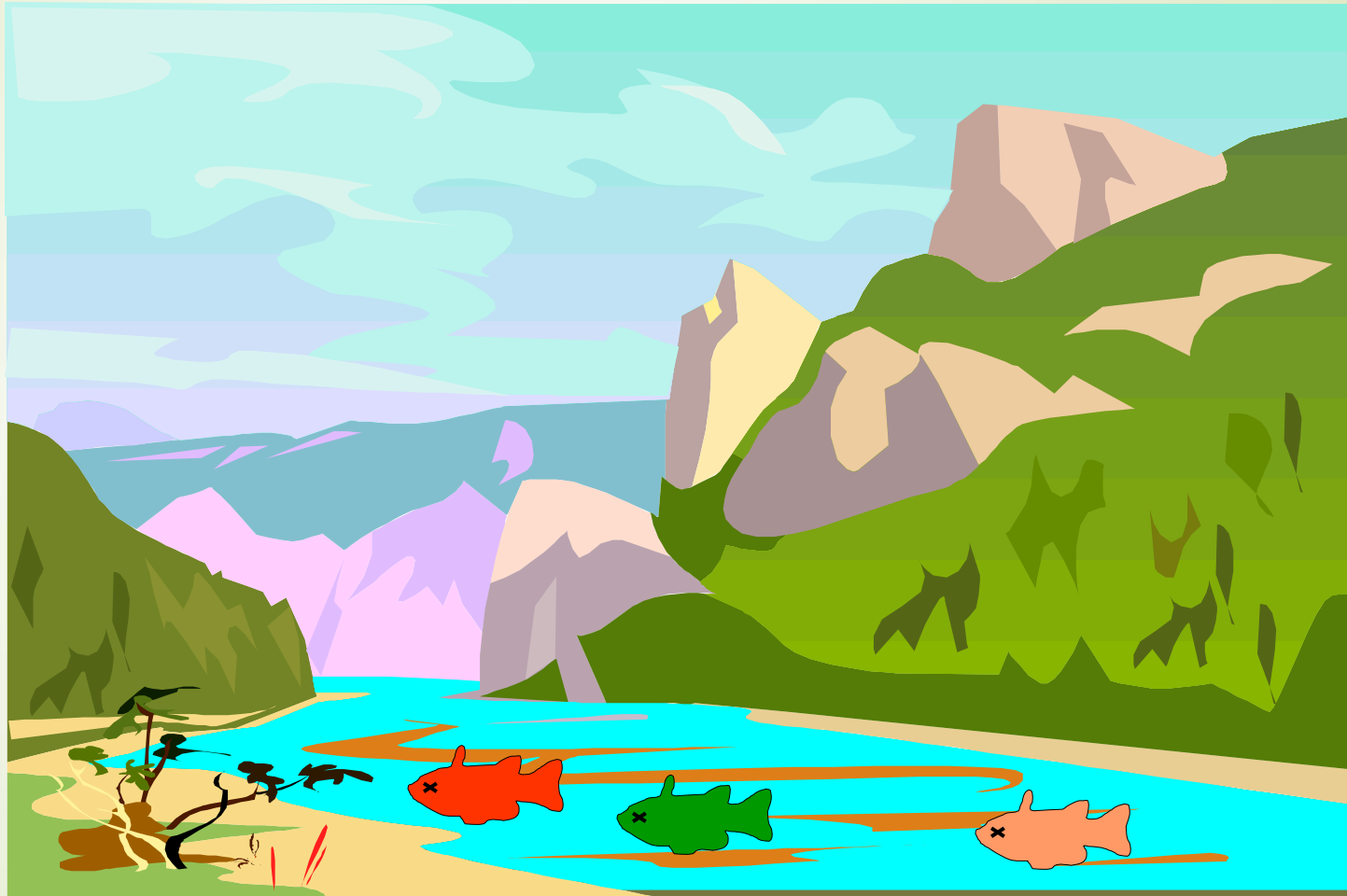
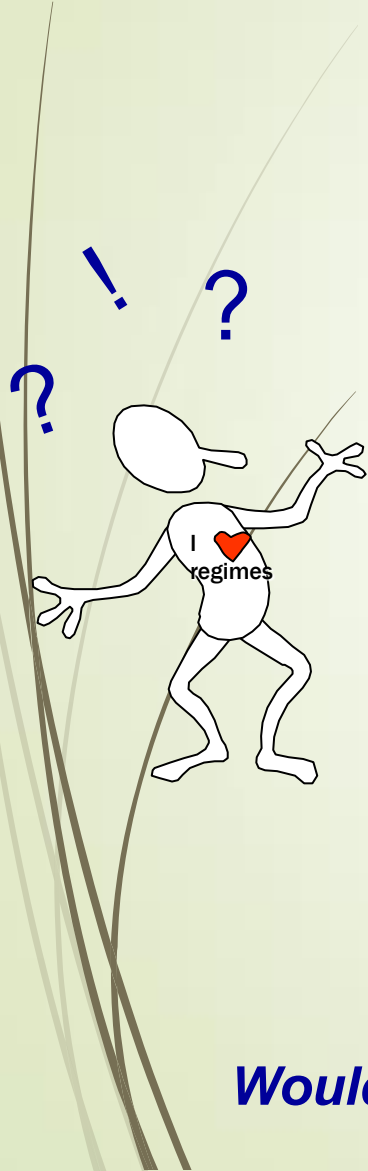
# The importance of identifying and structuring objectives common sticking point

## Confusing fundamental and mean objectives

Stated (fundamental) objective of stream fishery manager:  
*Natural Hydrologic Regime*

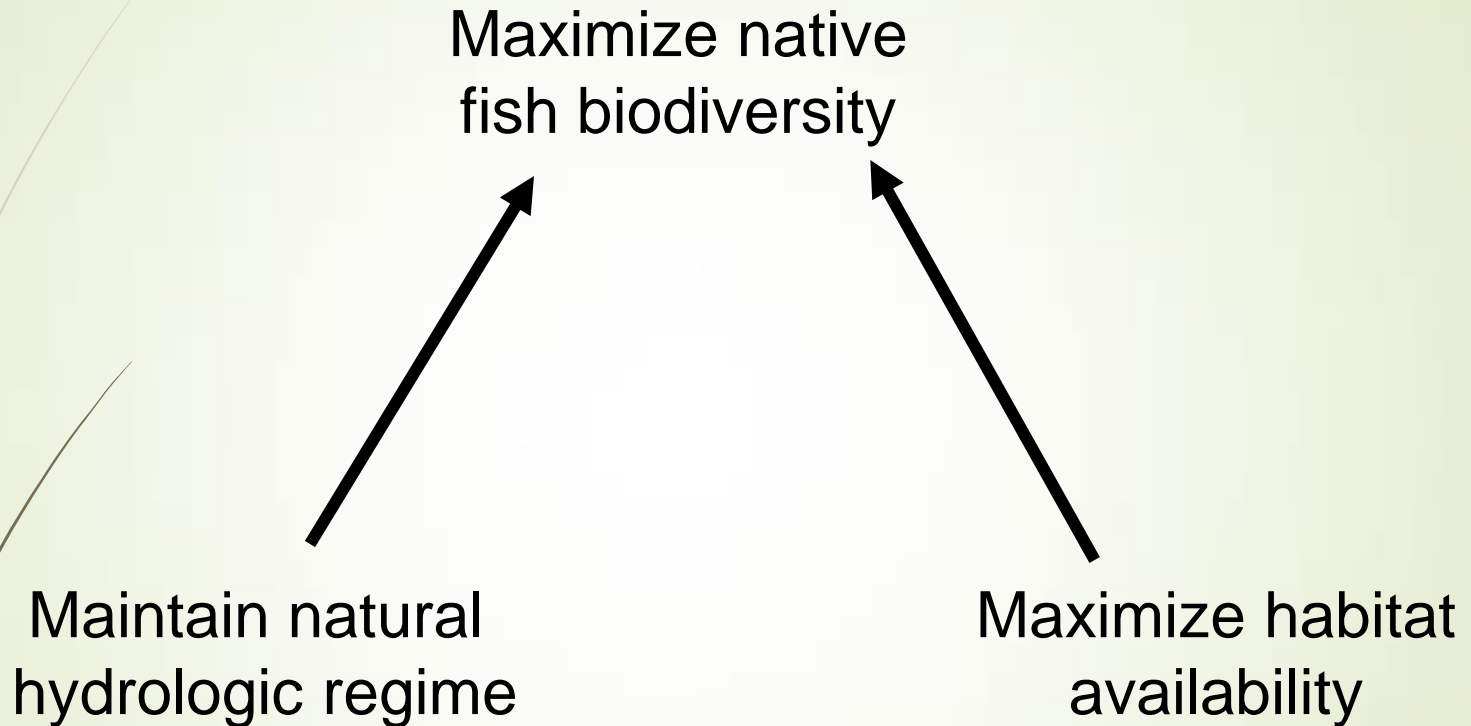


**Possible outcome:** The flow regime is natural but....  
*all the fish are dead*



***Would the fishery manager be happy with the outcome???***

# Means objectives (sometimes) help realize the fundamental objective



*Means objectives often are hypotheses about system dynamics*

# More common problems

Dismissing potential objectives due to perceived conflicts

Dismissing potential objectives due to perceived lack of information or complexity

*Values (objectives) masquerading as facts or process*



Please leave your  
model at the door

# Decision Support Process



Step 1: Identify the problem / decision situation

Step 2: Identify and Structure Objectives

Step 3: Identify decision alternatives

*These 3 steps = most difficult aspects of process*

# Step 4: Model building

Construct the model (decision support tool)

Simple (simple is good!)

Complex (can do but-- the rule of 6)

*But...Must include uncertainty*

Where do we get the information?

Empirical data

Published reports (meta-analysis)

“Expert” judgment



# Common forms of uncertainty in natural resources management decision making

Linguistic

Epistemic

Statistical uncertainty

Observational error

Structural uncertainty

Reducible

Aleatory

Environmental variability

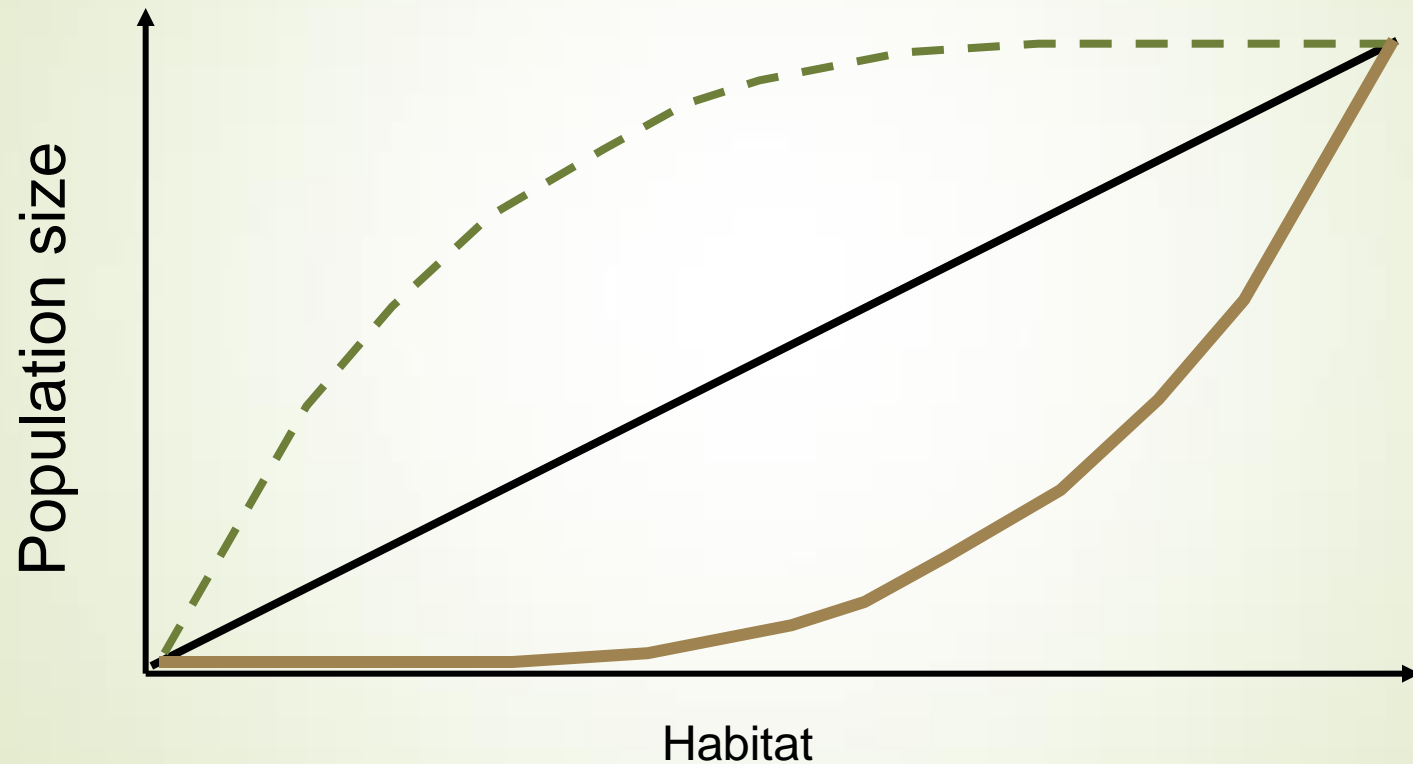
Demographic variability

Irreducible

# Often overlooked source of uncertainty

## Structural (System) uncertainty

due to incomplete understanding of system dynamics



*Incorporated into decision modeling using multiple models and model probabilities (weights)*

# **Step 5: Identify key uncertainties— *Sensitivity Analysis***

**An essential step**

**Basis for model simplification**

**Focus monitoring on decision-making**  
what do we need to know  
how much is enough

**Estimate value of information**  
collecting monitoring data  
more studies

# Example: Water availability for ecological needs in the ACF Basin

## Spatially explicit

Stream segment  
Flow, habitat, fish  
metapopulation models (43  
species)

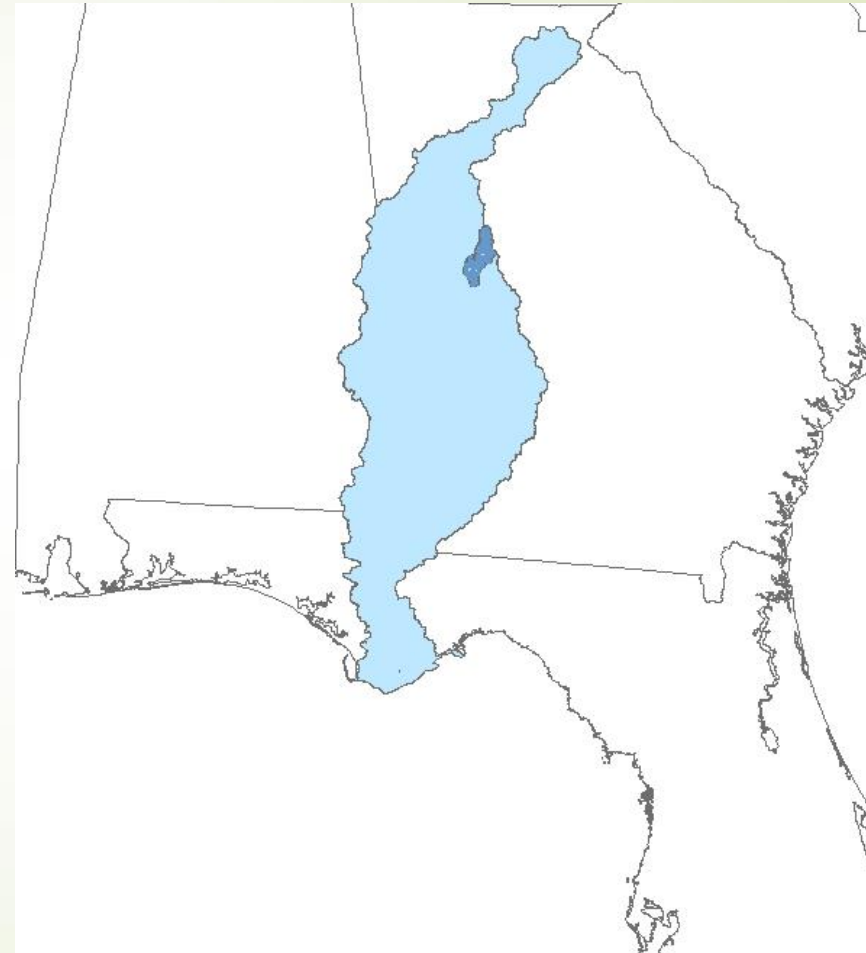
## Statistical uncertainty

Flow and habitat model  
errors

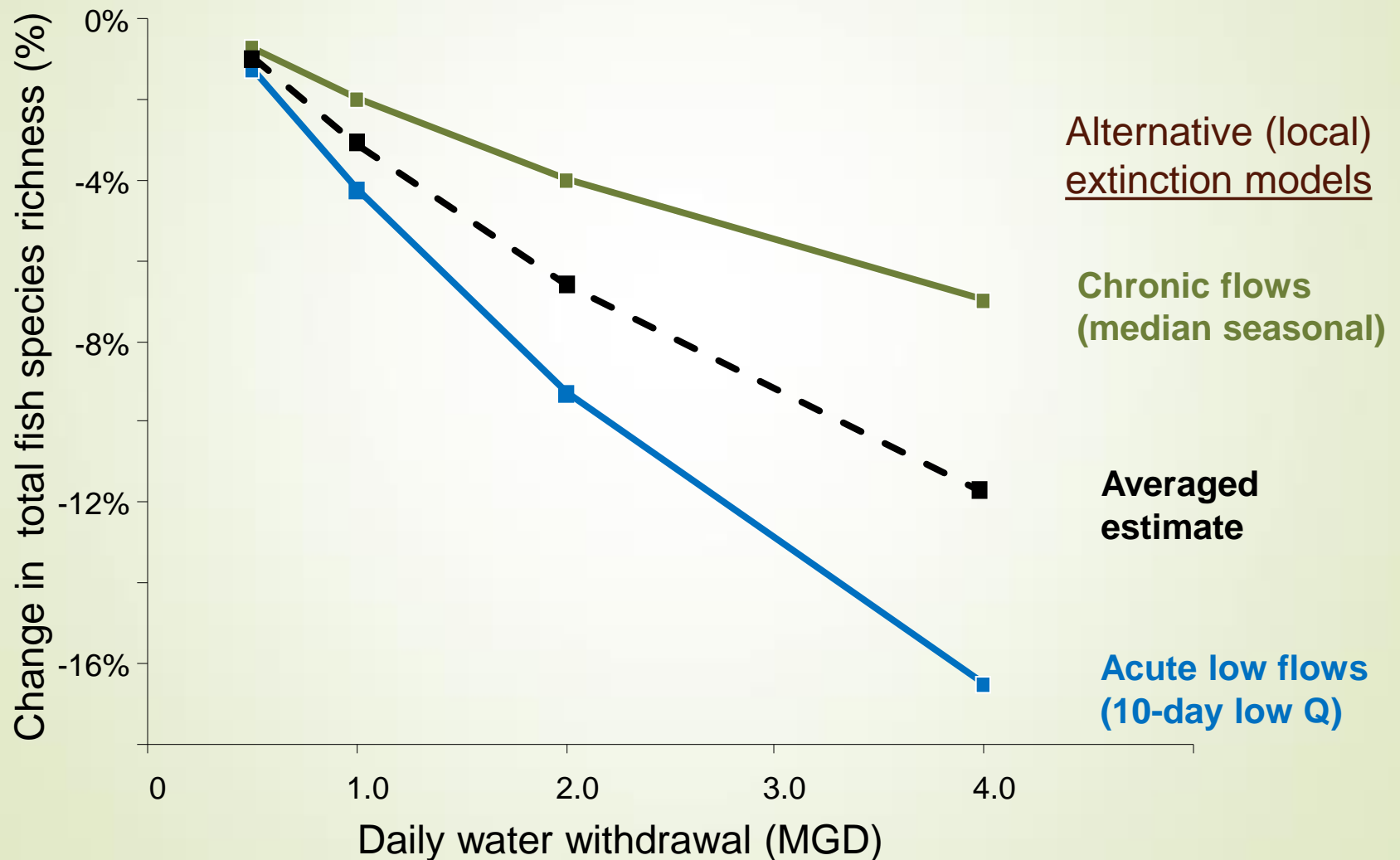
## Structural (system) uncertainty

Alternative fish population  
demographics models

Apalachicola, Chattahoochee, Flint Basin



# What assumptions/inputs affect the decision?



# Value of information

Expected value of decision if no uncertainty

- Model parameters

- Model inputs and system state

Currency that is valued by the decision-makers

- Fish population size

- Water available for use

- Others



# Value of imperfect information

## Value of sample information

Multi-species occupancy simulations

2 sample occasions, error (CV) ~ 35%

True richness, given estimated 25: 25 +/- 4

Value of sample information: **0.26 MGD**

4 sample occasions, error (CV) ~ 10%

True richness, given estimated 25: 25 +/- 2

Value of sample information: **0.49 MGD**

*Compare to EVPI = 0.61*

# Reducing Uncertainty

## Retrospective studies

Can provide a good initial basis for prediction

Usually confounded with other factors

## Experiments

Difficult to perform in many systems

Uncertainty reduction is *not* directed at resource objective (inefficient)

## Adaptive management

Can be done in virtually any resource system

No tradeoff necessary in resource objective

# Learning how a system works

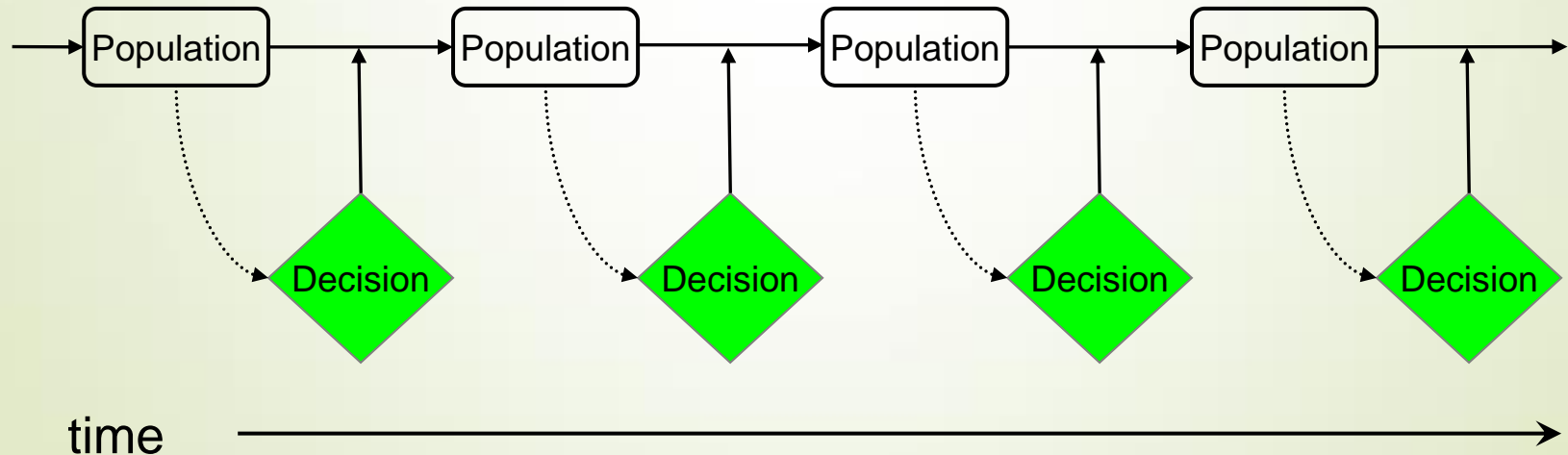
## Learn while managing (Adaptive Resource Management)

Decisions are made

Requires *sequential dynamic* decision-making: time and/or space

Probing (aka experimentation) not *necessary*

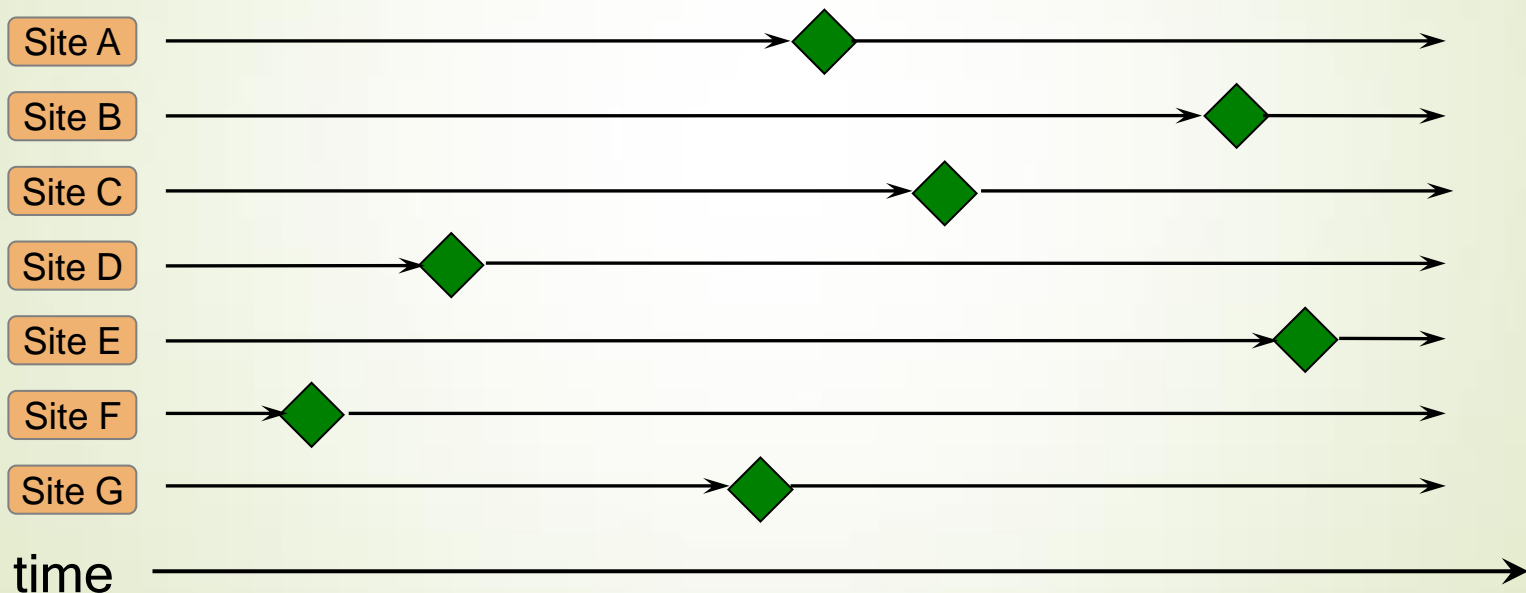
## Sequential decision-making through time



C. Moore

# Learning how a system works

## Sequential decision-making through space



C. Moore

# An illustration: Dueling professors

We have two potential management actions for swamp suckers:  
**increasing spawning habitat** or **increase juvenile rearing habitat**  
*(we don't have enough money to do both!)*

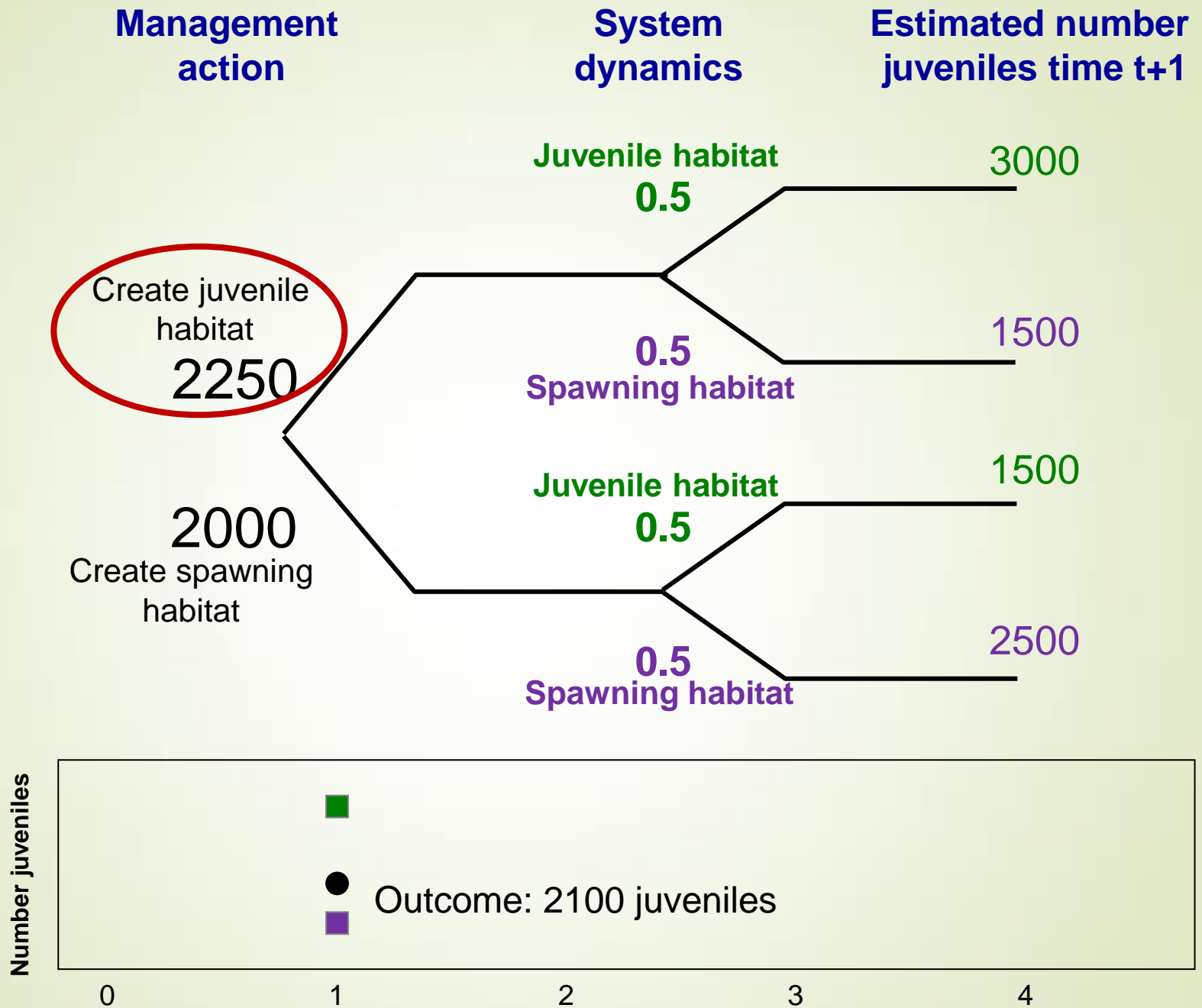


Professor Knowsittall, says that spawning habitat is limited. She estimates that there will be 2500 juvenile suckers produced if spawning habitat is increased, 1500 if not

Professor Nottoobrite thinks that juvenile rearing habitat is limited. He estimates that there will be 3000 juvenile suckers produced if rearing habitat is increased, 1500 if not



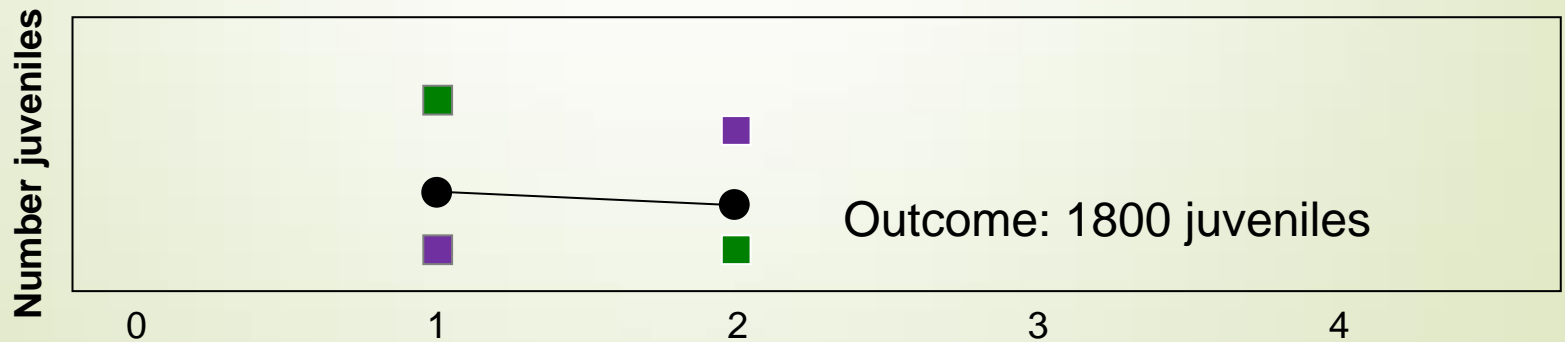
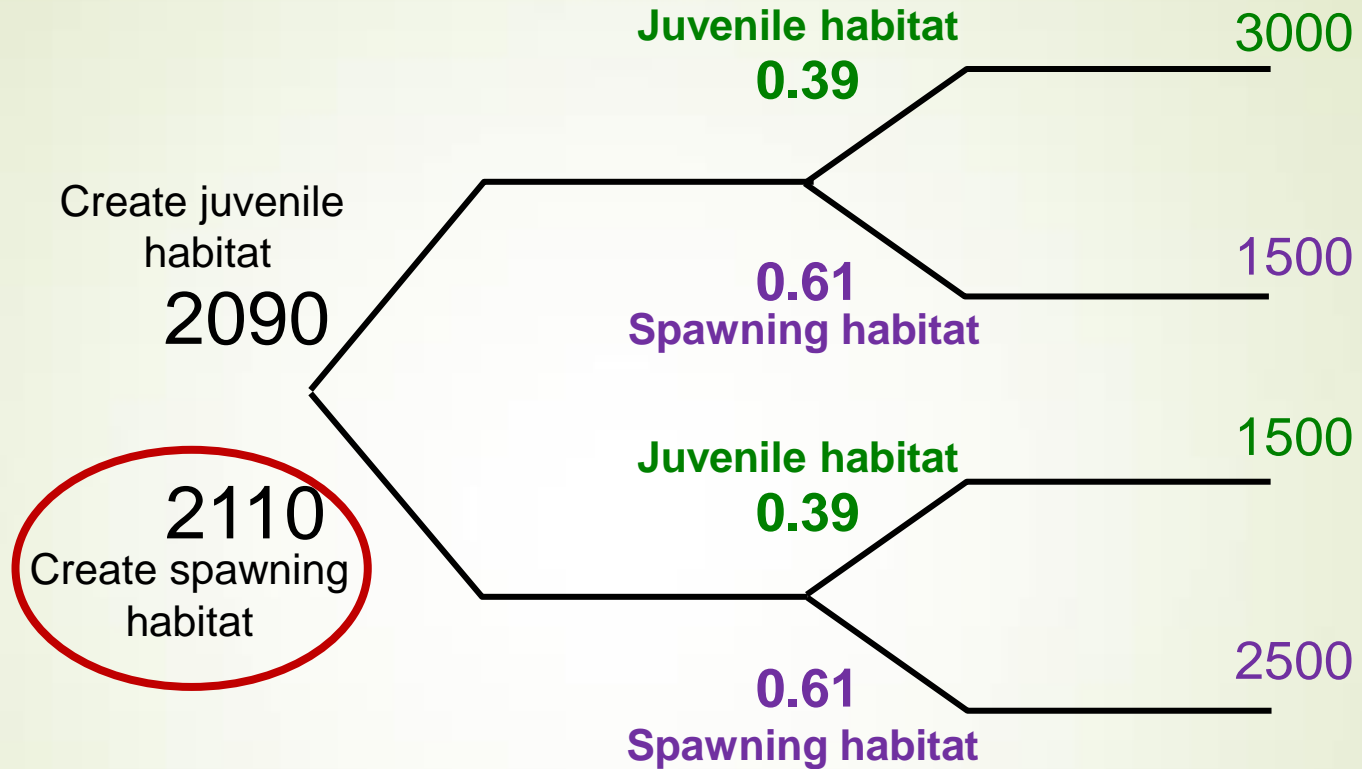
***What do we do?....***

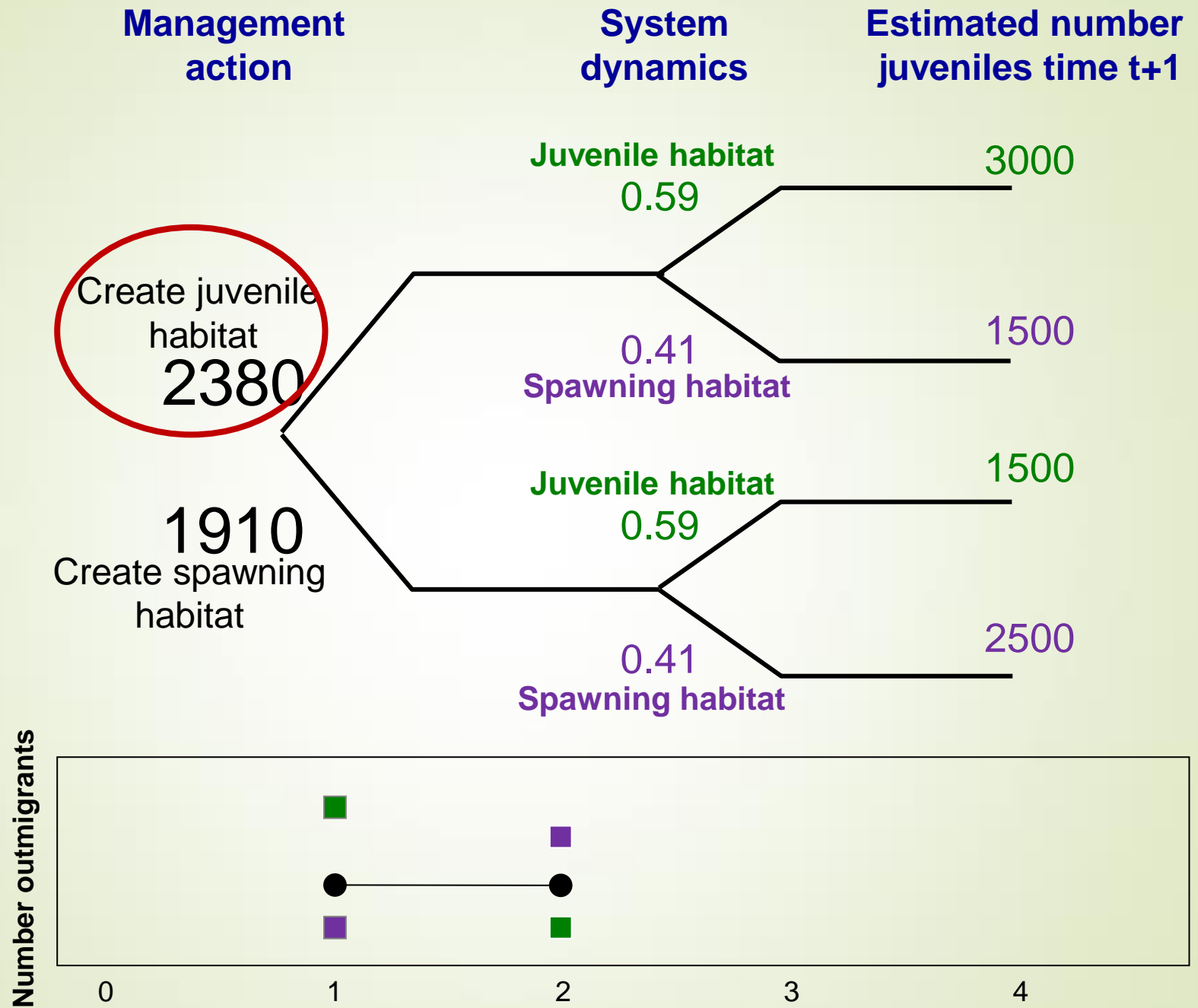


**Management  
action**

**System  
dynamics**

**Estimated number  
juveniles time t+1**





# Example: Reducing uncertainty

## ACF Basin: monitoring

**Spring and summer 2011- 2013**

**21 sites, 40- 100 m**

**Electrofishing and seining**

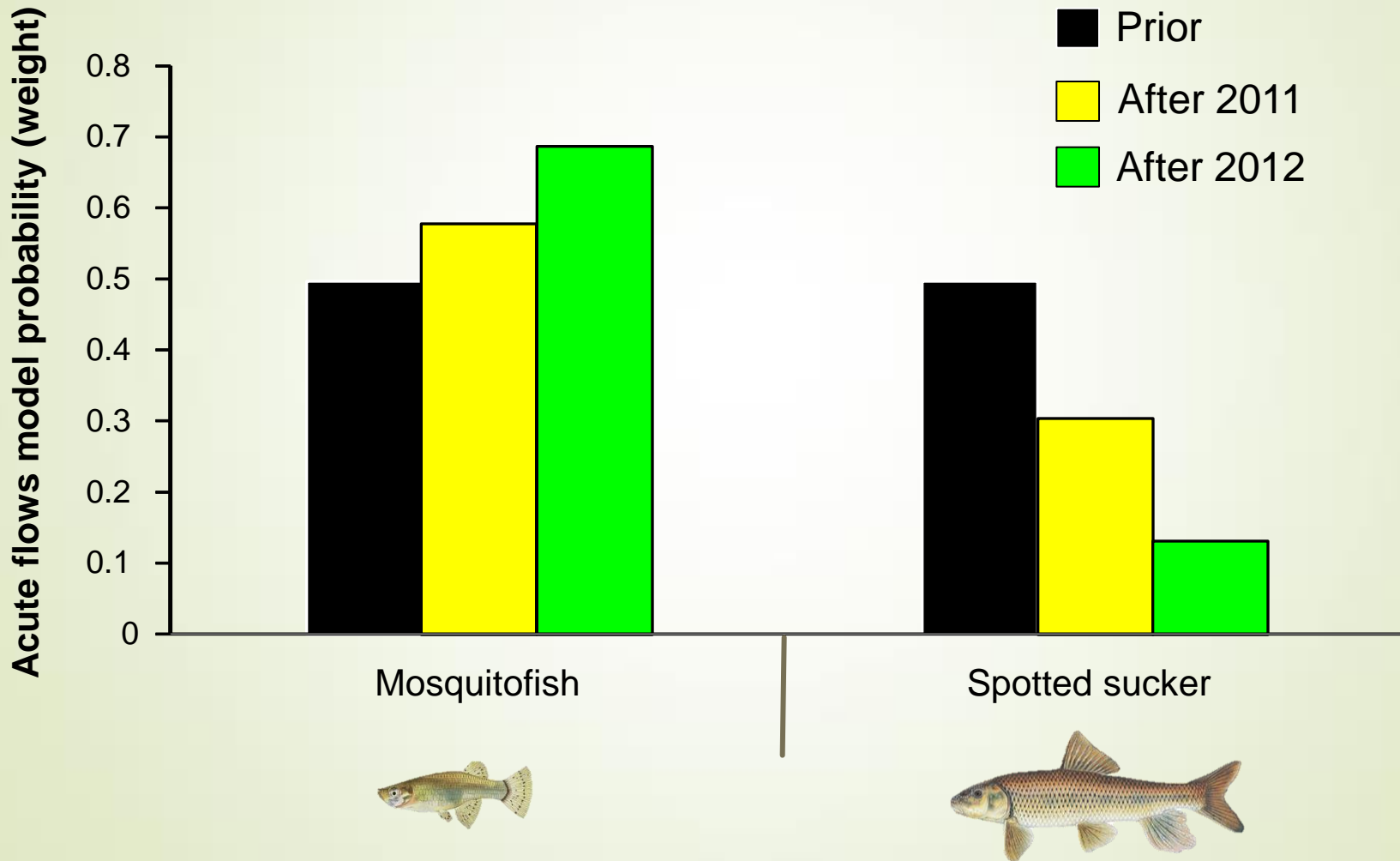
**Occupancy 2-3 visits season**



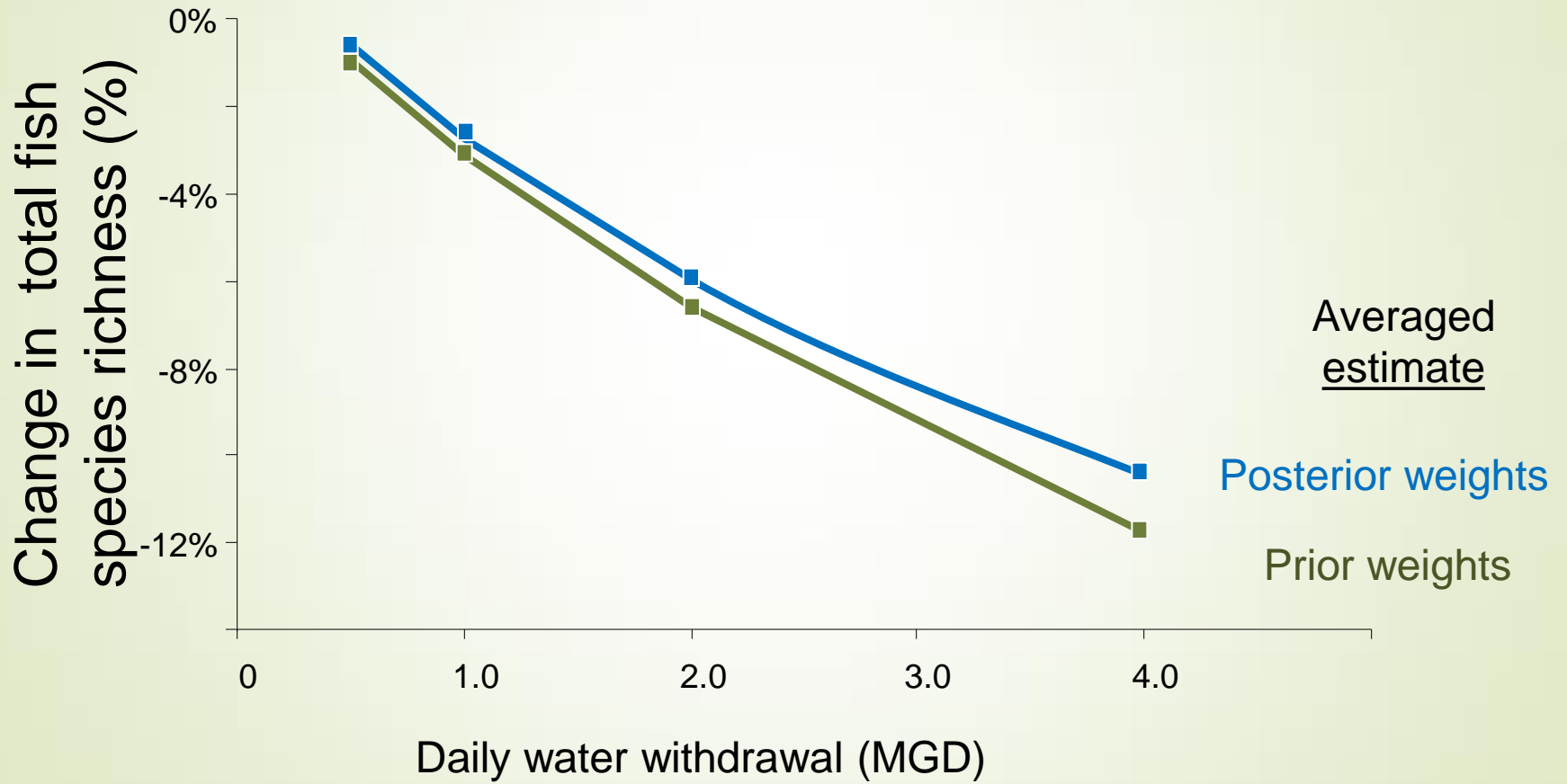
Flint River Basin



# Update model probabilities



# Updated estimates of water use effects



# Role of monitoring

## Traditional approach

Current state of the system

Trajectory (trends)

Integration of information *indirect*

Potential loss of information (knowledge)

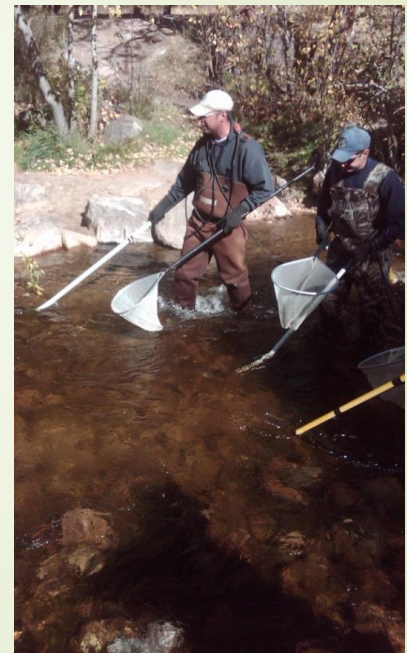
## Adaptive resource management (ARM)

Current state of the system

Information on system dynamics

Integration of information *direct*

Institutional memory contained in model(s)



# What if the models are wrong?

## Single and double loop learning

